

**LARGE DEVIATIONS AND THE INTERNAL FLUCTUATIONS
OF CRITICAL MEAN FIELD SYSTEMS**

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Received 22 June 1989

It is well known that the fluctuations of critical mean field systems are non-Gaussian. However, as shown in Papangelou (1989), the internal or second order fluctuations of critical Curie-Weiss models, i.e. fluctuations between subsystems, are Gaussian. In the present article we prove the same result by a large deviations approach which applies to more general mean field systems.

critical mean field fluctuations * large deviations * entropy * Laplace's method * Brownian bridge

1. Introduction

Systems of dependent random variables ('spins') $X_1^{(n)}, X_2^{(n)}, \dots, X_n^{(n)}$ with joint distributions of the form

$$P(X_1^{(n)} \in dx_1, X_2^{(n)} \in dx_2, \dots, X_n^{(n)} \in dx_n) \\ = \text{const.} \exp \left\{ \beta n G \left(\frac{x_1 + \dots + x_n}{n} \right) \right\} \prod_{i=1}^n \rho(dx_i)$$

have been considered extensively in the literature because of their significance for mean field models in statistical mechanics. One of the most interesting examples, the basic Curie-Weiss model, has $G(x) = \frac{1}{2}x^2$ and $\rho(\{1\}) = \rho(\{-1\}) = \frac{1}{2}$ [8, 15]. This and other examples exhibit 'phase transition' at a critical value β_c of β : for $\beta < \beta_c$ the asymptotic distribution of S_n/\sqrt{n} as $n \rightarrow \infty$, where S_n is the total field $X_1^{(n)} + \dots + X_n^{(n)}$, is Gaussian, while for $\beta = \beta_c$ a more severely scaled sum such as $S_n/n^{3/4}$, say, has a non-Gaussian asymptotic distribution; for $\beta > \beta_c$ the weak law of large numbers fails. (We assume for convenience that the mean of the probability measure ρ is 0.) The existence of a non-Gaussian limit in the critical case was first proved in [14] for the basic Curie-Weiss model and since then there have been many studies of this phenomenon, which is in fact a large deviations phenomenon, implicitly treated as such in [14], and explicitly so in [6].

This result on critical behaviour has been generalised in various directions [7, 11, 3, 4, 2] and has also been extended to mean field systems undergoing time evolution [5, 2]. Implicit in the treatment of the case $G(x) = \frac{1}{2}x^2$ in [7] is the fact that $\exp\{\frac{1}{2}x^2\}$ is the moment generating function of the standard normal distribution.

This was already exploited by Kac in [10] and leads in fact to a de Finetti representation of the joint distribution of $X_1^{(n)}, \dots, X_n^{(n)}$ as a mixture of i.i.d.'s. (This line of thought was pursued in [4], where the above result was extended to sequences of moment generating functions $\exp G_n(x)$.) However, the result is independent of this fact and, for instance, [11] and [3] rely exclusively on large deviations and Laplace's method.

For the case $G(x) = \frac{1}{2}x^2$ we showed in [12] that, despite the non-Gaussian nature of the critical fluctuations, the fluctuations between sub-blocks of spins are Gaussian. If, for example, the system consists of $2n$ random variables $X_1^{(2n)}, \dots, X_{2n}^{(2n)}$, then the asymptotic distribution of

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n X_i^{(2n)} - \frac{1}{\sqrt{n}} \sum_{i=n+1}^{2n} X_i^{(2n)}$$

is Gaussian. With this as a starting point we proved in [12] a number of limit theorems, of which the following is perhaps the most illuminating representative.

Define the standard polygonal line

$$\xi_n(t) = X_1^{(n)} + \dots + X_{[nt]}^{(n)} + (nt - [nt])X_{[nt]+1}^{(n)}, \quad 0 \leq t \leq 1.$$

Theorem 1 [12]. *Let $G(x) = \frac{1}{2}x^2$, $\beta = 1$ and suppose that the probability measure ρ has mean 0 and variance 1, and that*

$$\psi(s) := \log \int_{-\infty}^{\infty} e^{sx} \rho(dx) < \frac{1}{2}s^2$$

for all $s \neq 0$. Then there is (an even integer) $p > 2$ such that

$$\frac{\xi_n(t)}{n^{1-1/p}} = \frac{S_n}{n^{1-1/p}} t + \frac{1}{n^{1/2-1/p}} \eta_n(t), \quad 0 \leq t \leq 1,$$

where as $n \rightarrow \infty$ the random variable $S_n/n^{1-1/p}$ and the process $\eta_n(t)$, $0 \leq t \leq 1$, converge jointly in distribution, the former to a random variable Y with probability density function $\text{const. exp}(-\lambda y^p)$, $-\infty < y < \infty$, and the latter to a Brownian bridge independent of Y . \square

Another theorem, which we will not repeat here, states that a differently scaled polygonal line is, asymptotically, the sum $Yt + \zeta(t)$ of a random straight line Yt as in Theorem 1 and an independent standard Brownian motion $\zeta(t)$. We refer the reader to [12] for further discussion of these results.

The proofs given in [12] relied on the de Finetti representation mentioned above. In the present paper we give, as promised in [12], an alternative approach which relies exclusively on large deviations techniques and requires no assumption that $\exp G(x)$ is a moment generating function. We will not re-establish here analogues of all the limit theorems stated in [12]; we will only prove the convergence of finite-dimensional distributions behind the assertion of Theorem 1. The result

obtained can be described informally as follows. Suppose the system consists of kn spins grouped into k subsets (k fixed) of n spins each, and let $S_n^1, S_n^2, \dots, S_n^k$ be the block spins of these subsets (S_n^j is $\sum_i X_i^{(nk)}$ with the summation restricted to the j th subset). Under the hypotheses set out below there is $p > 2$ such that, if the linear space \mathbb{R}^k is scaled by the factor $n^{-(1-1/p)}$ in the direction of the principal diagonal vector $(1, 1, \dots, 1)$ and by the factor $n^{-1/2}$ in all directions orthogonal to the diagonal, then the distribution of the random vector $(S_n^1, S_n^2, \dots, S_n^k)$ in \mathbb{R}^k converges, as $n \rightarrow \infty$, to the product of a non-Gaussian distribution along the principal diagonal and a standard rotationally invariant Gaussian distribution on the $(k-1)$ -dimensional linear subspace orthogonal to the diagonal.

The proofs given below require nothing more than Cramér's original one-dimensional large deviations estimates for tail probabilities: no local large deviations theorem will be used. Prior to proving the main result (Theorems 3 and 4) we tread in Section 2 on the familiar territory of the Laplace approach (cf. [11, 3]), albeit along a somewhat novel path, culminating in the statement of the non-Gaussian fluctuations of the field. We have two purposes in doing this. The first is to state and prove on the way the new Proposition 3 and its corollary, which are essential for Section 3. The second is to show how all proofs in Sections 2 and 3 can be based on Proposition 1 and the quite weak assumptions made in Section 2. These assumptions cover cases where $G(x)$ is genuinely quadratic throughout its domain.

I take this opportunity to thank Gérard Ben Arous for making available to me a preprint of [2] and for a useful conversation on the subject.

2. The fluctuations of the field

Throughout the paper we assume that ρ is a probability measure on the Borel σ -field of the real line, with mean 0 and variance 1 and such that $\int_{-\infty}^{\infty} e^{sx} \rho(dx) < \infty$ for all $s \in (-\infty, \infty)$. The cumulant generating function of ρ ,

$$\psi(s) = \log \int_{-\infty}^{\infty} e^{sx} \rho(dx)$$

is then convex and real analytic on $(-\infty, \infty)$ and satisfies $\psi'(0) = 0$, $\psi''(0) = 1$. The entropy function of ρ is

$$H(x) = \sup_t \{tx - \psi(t)\}.$$

If ζ denotes the supremum and θ the infimum of the support of ρ ($-\infty \leq \theta < 0 < \zeta \leq \infty$), then H is convex and real analytic in (θ, ζ) and

$$H(x) = \frac{1}{2}x^2 + O(x^3) \quad \text{as } x \rightarrow 0.$$

The derivative ψ' is a strictly increasing mapping of $(-\infty, \infty)$ on to (θ, ζ) and if, given $x \in (\theta, \zeta)$, we let s be the unique real number such that $\psi'(s) = x$, then $H(x) = s\psi'(s) - \psi(s)$. The inverse of $x = \psi'(s)$ is $s = H'(x)$ and it can be shown that

$H'(x) \rightarrow \infty$ as $x \rightarrow \zeta$ and $H(x)/x \rightarrow \infty$ as $x \rightarrow \infty$. Analogously for negative x . We refer the reader to [1] and [8] for basic facts.

All our proofs below will be based entirely on three classical facts about one-dimensional large deviations, going back to Cramér's work. We collect these facts in the following proposition.

Denote by F_n the (right-continuous) distribution function of $(1/n) \sum_{i=1}^n W_i$ where W_1, \dots, W_n are i.i.d. random variables with distribution ρ .

Proposition 1. (i) For all $x \geq 0$,

$$1 - F_n(x) \leq \exp\{-nH(x)\}. \quad (1)$$

(ii) If $0 < b < \zeta$, then there exists a constant $C(b)$ such that for all $x \in [0, b]$,

$$x(1 - F_n(x)) \leq \frac{C(b)}{\sqrt{n}} \exp\{-nH(x)\}. \quad (2)$$

(iii) Let $p > 2$ and $0 < a < b < \infty$. If $n \rightarrow \infty$ then, for $an^{-1/p} \leq x \leq bn^{-1/p}$,

$$x(1 - F_n(x)) = \frac{1 + o(1)}{\sqrt{2\pi n}} \exp\{-nH(x)\} \quad (3)$$

where $o(1) \rightarrow 0$ uniformly for x in the above range. \square

Analogous statements can of course be made for negative x .

The first statement is an immediate and familiar consequence of the definition of H , since

$$1 - F_n(x) \leq \exp(-tnx) E(e^{t \sum_{i=1}^n W_i}) = \exp\{-n(tx - \psi(t))\}$$

for arbitrary t . The second and third statements follow from the considerations of [9, pp. 548–553]. See also [13, Chapter VIII].

In what follows we will be concerned with a system of random variables $X_1^{(n)}, X_2^{(n)}, \dots, X_n^{(n)}$ having joint distribution

$$\begin{aligned} P(X_1^{(n)} \in dx_1, X_2^{(n)} \in dx_2, \dots, X_n^{(n)} \in dx_n) \\ = \mathcal{Z}_n^{-1} \exp\left\{nG\left(\frac{x_1 + \dots + x_n}{n}\right)\right\} \prod_{i=1}^n \rho(dx_i), \end{aligned} \quad (4)$$

where $G(x)$ is defined and finite for all x and \mathcal{Z}_n is the appropriate norming constant. We make the following assumptions about G .

(I) G is everywhere differentiable, its derivative G' is bounded on bounded sets and $G'(x) = x(1 + o(1))$ as $x \rightarrow 0$.

(II) $G(x) < H(x)$ for all $x \neq 0$.

(III) $\limsup G(x)/H(x) < 1$ as $x \rightarrow \zeta$ or $x \rightarrow \theta$.

(IV) There exist $\lambda > 0$ and $p > 2$ such that $G(x) - H(x) = -\lambda|x|^p + o(|x|^p)$ as $x \rightarrow 0$.

Note that (I) implies $G(x) = \frac{1}{2}x^2 + o(x^2)$ as $x \rightarrow 0$ and that (III) is trivially true if $\zeta < \infty$ and $\theta > -\infty$. Condition (III) is sufficient to guarantee that

$\int_{-\infty}^{\infty} \exp G(x) \rho(dx) < \infty$ and hence that $\mathcal{X}_n < \infty$ for all n , even if $\zeta = \infty$ or $\theta = \infty$, as will be shown below. It is also less restrictive than the condition “ $|G(x)| \leq a|x| + b$ for some a, b and all sufficiently large $|x|$ ”, assumed in [11] and [3], since the latter excludes the function $G(x) = \frac{1}{2}x^2$ if $\zeta = \infty$ or $\theta = -\infty$.

Lemma 1. *In order that (I), (II) and (III) hold, it is sufficient that G' exist,*

$$0 < G'(\psi'(s))/s < 1 \quad \text{for } s \neq 0$$

and

$$\limsup G'(\psi'(s))/s < 1 \quad \text{as } s \rightarrow \pm\infty. \quad \square$$

In fact the given conditions imply, for instance, that $G'(x) < H'(x)$ for all $x \in (0, \zeta)$ and that there exist $\delta > 0$ and $x_0 \in (0, \zeta)$ such that $G'(x) < (1 - \delta)H'(x)$ for $x \in (x_0, \zeta)$. It follows that $G(x) - H(x)$ is decreasing for $x > 0$ and that $G(x) \leq (1 - \delta)H(x) + G(x_0) - (1 - \delta)H(x_0)$ for $x \geq x_0$. The latter implies (III) if $\zeta = \infty$, since $H(x)/x \rightarrow \infty$ as $x \rightarrow \infty$.

As an example, if $G(x) = \frac{1}{2}x^2$ and if $\psi'''(s) < 0$ for $s > 0$ and $\psi'''(s) > 0$ for $s < 0$ (GHS inequality, cf. [7, p. 137]), then (I), (II), (III) hold.

If $S_n = X_1^{(n)} + X_2^{(n)} + \cdots + X_n^{(n)}$, then the distribution of S_n/n is given by

$$P(\alpha < S_n/n \leq \beta) = \frac{1}{\mathcal{X}_n} \int_{\alpha}^{\beta} \exp\{nG(x)\} dF_n(x),$$

where we use the convention that \int_{α}^{β} denotes $\int_{[\alpha, \beta]}$. If $0 \leq \alpha < \beta < \infty$, an integration by parts shows that

$$\begin{aligned} & \int_{\alpha}^{\beta} \exp\{nG(x)\} dF_n(x) \\ &= n \int_{\alpha}^{\beta} \exp\{nG(x)\} G'(x) (1 - F_n(x)) dx \\ & \quad + (1 - F_n(\alpha)) \exp\{nG(\alpha)\} - (1 - F_n(\beta)) \exp\{nG(\beta)\}, \end{aligned} \quad (5)$$

where by (1) and (II) each of the last two terms is bounded by 1.

The propositions which follow are formulated for positive a, b but it should be obvious that analogous statements hold for negative values.

Proposition 2. *For $0 \leq a < b < \infty$,*

$$\lim_{n \rightarrow \infty} n^{-1/2+1/p} \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{nG(x)\} dF_n(x) = \frac{1}{\sqrt{2\pi}} \int_a^b \exp(-\lambda z^p) dz.$$

Proof. By (5), (1) and (II),

$$\begin{aligned} & \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{nG(x)\} dF_n(x) \\ &= n \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{nG(x)\} G'(x) (1 - F_n(x)) dx + O(1). \end{aligned}$$

If $a > 0$, then by (2) the last term is in fact $o(1)$ and (I), (3) and (IV) imply that the right-hand side is equal to

$$\begin{aligned} & \frac{1+o(1)}{\sqrt{2\pi}} n^{1/2} \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{n(G(x) - H(x))\} dx + o(1) \\ &= \frac{1+o(1)}{\sqrt{2\pi}} n^{1/2} \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{n(-\lambda x^p + o(x^p))\} dx + o(1) \\ &= \frac{1+o(1)}{\sqrt{2\pi}} n^{1/2-1/p} \int_a^b \exp\left\{-\lambda z^p + no\left(\frac{z^p}{n}\right)\right\} dz + o(1) \end{aligned}$$

which proves the result for $a > 0$. Now notice that by (I), (2) and (II),

$$\begin{aligned} & n \int_0^{an^{-1/p}} \exp\{nG(x)\} G'(x)(1 - F_n(x)) dx \\ & \leq Cn^{1/2} \int_0^{an^{-1/p}} \exp\{n(G(x) - H(x))\} dx \leq Can^{1/2-1/p} \end{aligned}$$

for some constant C , and by taking a sufficiently small a we can extend the result to the case where the lower limit of integration is 0. \square

Proposition 3. *If $a_n \rightarrow a > 0$ as $n \rightarrow \infty$, then for any finite $u < v$,*

$$\lim_{n \rightarrow \infty} \int_{a_n n^{-1/p} + un^{-1/2}}^{a_n n^{-1/p} + vn^{-1/2}} \exp\{nG(x)\} dF_n(x) = \frac{v-u}{\sqrt{2\pi}} \exp(-\lambda a^p).$$

Proof. Just as in the proof of Proposition 2, we see that the integral on the left is equal to

$$\begin{aligned} & \frac{1+o(1)}{\sqrt{2\pi}} n^{1/2-1/p} \int_{a_n + un^{1/2+1/p}}^{a_n + vn^{1/2+1/p}} \exp\left\{-\lambda z^p + no\left(\frac{z^p}{n}\right)\right\} dz + o(1) \\ &= (1+o(1)) \frac{v-u}{\sqrt{2\pi}} [(v-u)n^{-1/2+1/p}]^{-1} \int_{a_n + un^{-1/2+1/p}}^{a_n + vn^{-1/2+1/p}} \exp(-\lambda z^p) dz + o(1) \end{aligned}$$

from which the result follows immediately. \square

We will find it convenient to state Proposition 3 in slightly different notation. First notice that an integral such as $\int_a^\beta \exp\{nG(x)\} dF_n(x)$ can also be written as

$$\int_{\alpha\sqrt{n}}^{\beta\sqrt{n}} \exp\left\{nG\left(\frac{x}{\sqrt{n}}\right)\right\} dF_n\left(\frac{x}{\sqrt{n}}\right),$$

where $F_n(x/\sqrt{n})$ as a function of x is the distribution function of $(1/\sqrt{n}) \sum_{i=1}^n W_i$. (See the paragraph preceding Proposition 1.) Proposition 3 then asserts that

$$\lim_{n \rightarrow \infty} \int_{a_n n^{1/2-1/p} + u}^{a_n n^{1/2-1/p} + v} \exp\left\{nG\left(\frac{x}{\sqrt{n}}\right)\right\} dF_n\left(\frac{x}{\sqrt{n}}\right) = \frac{1}{\sqrt{2\pi}} \exp(-\lambda a^p)(v-u).$$

If we define a sequence of measures μ_n , $n = 1, 2, \dots$, on the Borel σ -field \mathcal{B} of \mathbb{R} by setting for $B \in \mathcal{B}$,

$$\mu_n(B) = \int_B \exp\left\{nG\left(\frac{x}{\sqrt{n}}\right)\right\} dF_n\left(\frac{x}{\sqrt{n}}\right), \quad (6)$$

we see that Proposition 3 asserts the convergence of the measures $\mu_n(a_n n^{1/2-1/p} + B)$ (where $x + B$ is the obvious translate of B) to a constant multiple of Lebesgue measure. Denoting measures by their differentials, we can state this as follows.

Corollary 1. *If $a_n \rightarrow a > 0$ then, on any bounded interval, the sequence of measures*

$$\mu_n(a_n n^{1/2-1/p} + dt)$$

converges weakly to the measure

$$\frac{1}{\sqrt{2\pi}} \exp(-\lambda a^p) dt. \quad \square$$

This can also be stated as follows: if $a_n \rightarrow a > 0$, then the measure $\mathcal{L}_n P((S_n/\sqrt{n} - a_n n^{1/2-1/p}) \in dt)$ converges weakly to $(1/\sqrt{2\pi}) \exp(-\lambda a^p) dt$ on every bounded interval.

Next we turn to the tails of our integrals.

Proposition 4. *There is a constant K such that for all $b > 0$,*

$$\limsup_{n \rightarrow \infty} n^{-1/2+1/p} \int_{bn^{-1/p}}^{\infty} \exp\{nG(x)\} dF_n(x) \leq K \int_b^{\infty} \exp(-\tfrac{1}{2}\lambda z^p) dz.$$

Proof. Choose $\varepsilon > 0$ so small that $G(x) - H(x) < -\frac{1}{2}\lambda x^p$ for $0 \leq x \leq \varepsilon$ and split the integral on the left into

$$\int_{bn^{-1/p}}^{\varepsilon} + \int_{\varepsilon}^{\infty} := I_n + J_n.$$

Consider I_n first. By (5) it is sufficient to consider

$$I'_n = n \int_{bn^{-1/p}}^{\varepsilon} \exp\{nG(x)\} G'(x)(1 - F_n(x)) dx.$$

By (1) and (2) there is a constant K such that

$$\begin{aligned} |I'_n| &\leq K n^{1/2} \int_{bn^{-1/p}}^{\varepsilon} \exp\{n(G(x) - H(x))\} dx \\ &\leq K n^{1/2} \int_{bn^{-1/p}}^{\varepsilon} \exp\{-\tfrac{1}{2}n\lambda x^p\} dx \\ &\leq K n^{1/2-1/p} \int_b^{\varepsilon n^{1/p}} \exp(-\tfrac{1}{2}\lambda z^p) dz. \end{aligned} \quad (7)$$

Next consider J_n . By (III) there is $\delta > 0$ such that

$$\begin{aligned} J_n &\leq \int_{\varepsilon}^{\xi} \exp\{n(1-\delta)H(x)\} dF_n(x) \\ &= n(1-\delta) \int_{\varepsilon}^{\xi} \exp\{n(1-\delta)H(x)\} H'(x)(1-F_n(x)) dx \\ &\quad + (1-F_n(\varepsilon)) \exp\{n(1-\delta)H(\varepsilon)\}. \end{aligned}$$

By (1) the second term on the right is at most $\exp\{-n\delta H(\varepsilon)\}$ while the first term on the right is at most

$$n(1-\delta) \int_{\varepsilon}^{\xi} \exp\{-n\delta H(x)\} H'(x) dx = \frac{1-\delta}{\delta} [e^{-n\delta H(\varepsilon)} - e^{-n\delta H(\xi)}].$$

This and (7) imply the proposition. \square

We can now deduce that the familiar statement on the non-Gaussian fluctuations of the field, referred to in the introduction, can be made under the relatively weak assumptions of this section.

Theorem 2.

$$(i) \quad \lim_{n \rightarrow \infty} \mathcal{L}_n / n^{1/2-1/p} = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \exp(-\lambda |x|^p) dx.$$

(ii) For $-\infty \leq a < b \leq \infty$,

$$\lim_{n \rightarrow \infty} P(a < S_n / n^{1-1/p} \leq b) = \int_a^b \exp(-\lambda |x|^p) dx / \int_{-\infty}^{\infty} \exp(-\lambda |x|^p) dx.$$

\square

Both parts follow from Propositions 2 and 4 and the fact that

$$\begin{aligned} P(a < S_n / n^{1-1/p} \leq b) &= P(an^{-1/p} < S_n / n \leq bn^{-1/p}) \\ &= \frac{1}{\mathcal{L}_n} \int_{an^{-1/p}}^{bn^{-1/p}} \exp\{nG(x)\} dF_n(x). \end{aligned} \quad (8)$$

If we denote by $Q_n(dx)$ the distribution of $S_n / n^{1-1/p}$,

$$Q_n(dx) = P(S_n / n^{1-1/p} \in dx), \quad (9)$$

then the second part of the theorem may be formulated as follows.

Corollary 2. As $n \rightarrow \infty$, the sequence of distributions $Q_n(dx)$ $n = 1, 2, \dots$, converges weakly to the distribution $\text{const.} \exp(-\lambda |x|^p) dx$. \square

3. The internal fluctuations

Let k be a fixed positive integer and consider kn random variables $X_1^{(kn)}, X_2^{(kn)}, \dots, X_{kn}^{(kn)}$ with joint distribution given by (4) but with n replaced by

kn . Define

$$S_n^j = X_{(j-1)n+1}^{(kn)} + \cdots + X_{jn}^{(kn)}, \quad j = 1, 2, \dots, k.$$

The joint distribution of $S_n^1, S_n^2, \dots, S_n^k$ is given by

$$\begin{aligned} P\left(\frac{S_n^1}{\sqrt{n}} \in dz_1, \frac{S_n^2}{\sqrt{n}} \in dz_2, \dots, \frac{S_n^k}{\sqrt{n}} \in dz_k\right) \\ = \frac{1}{\mathcal{L}_{kn}} \exp\left\{knG\left(\frac{z_1 + \cdots + z_k}{k\sqrt{n}}\right)\right\} \prod_{j=1}^k dF_n\left(\frac{z_j}{\sqrt{n}}\right). \end{aligned}$$

In the present section we are interested in the joint distribution of

$$\frac{S_n^k}{n^{1-1/p}}, \quad \frac{S_n^1}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}}, \frac{S_n^2}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}}, \dots, \frac{S_n^{k-1}}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}}.$$

Let $0 < a < b$, $u_1 < v_1$, $u_2 < v_2$, \dots , $u_{k-1} < v_{k-1}$; then

$$\begin{aligned} P\left(a < \frac{S_n^k}{n^{1-1/p}} \leq b, u_1 < \frac{S_n^1}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}} \leq v_1, \dots, u_{k-1} < \frac{S_n^{k-1}}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}} \leq v_{k-1}\right) \\ = P\left(an^{1/2-1/p} < \frac{S_n^k}{\sqrt{n}} \leq bn^{1/2-1/p}, \frac{S_n^k}{\sqrt{n}} + u_1 < \frac{S_n^1}{\sqrt{n}} \leq \frac{S_n^k}{\sqrt{n}} + v_1, \right. \\ \left. \dots, \frac{S_n^k}{\sqrt{n}} + u_{k-1} < \frac{S_n^{k-1}}{\sqrt{n}} \leq \frac{S_n^k}{\sqrt{n}} + v_{k-1}\right) \\ = \frac{1}{\mathcal{L}_{kn}} \int_{an^{1/2-1/p}}^{bn^{1/2-1/p}} \left[\int_{z_k+u_1}^{z_k+v_1} \cdots \int_{z_k+u_{k-1}}^{z_k+v_{k-1}} \exp\left\{knG\left(\frac{z_1 + \cdots + z_k}{k\sqrt{n}}\right)\right\} \right. \\ \left. \cdot \prod_{j=1}^{k-1} dF_n\left(\frac{z_j}{\sqrt{n}}\right) \right] dF_n\left(\frac{z_k}{\sqrt{n}}\right). \end{aligned}$$

In terms of the notation of (6) this is

$$\begin{aligned} = \frac{1}{\mathcal{L}_{kn}} \int_{an^{1/2-1/p}}^{bn^{1/2-1/p}} \left[\int_{z_k+u_1}^{z_k+v_1} \cdots \int_{z_k+u_{k-1}}^{z_k+v_{k-1}} \exp\left\{knG\left(\frac{z_1 + \cdots + z_k}{k\sqrt{n}}\right) \right. \right. \\ \left. \left. - nG\left(\frac{z_1}{\sqrt{n}}\right) - \cdots - nG\left(\frac{z_k}{\sqrt{n}}\right)\right\} \prod_{j=1}^{k-1} \mu_n(dz_j) \right] \\ \cdot \exp\left\{nG\left(\frac{z_k}{\sqrt{n}}\right)\right\} dF_n\left(\frac{z_k}{\sqrt{n}}\right). \end{aligned} \quad (10)$$

Keeping z_k fixed, change the variables z_1, \dots, z_{k-1} in the multiple integral in square brackets as follows:

$$z_1 = z_k + t_1, \quad z_2 = z_k + t_2, \quad \dots, \quad z_{k-1} = z_k + t_{k-1}.$$

The integral in square brackets in (10) is then seen to be equal to

$$\int_{u_1}^{v_1} \cdots \int_{u_{k-1}}^{v_{k-1}} \exp \left\{ knG \left(\frac{z_k}{\sqrt{n}} + \frac{t_1 + \cdots + t_{k-1}}{k\sqrt{n}} \right) - nG \left(\frac{z_k}{\sqrt{n}} + \frac{t_1}{\sqrt{n}} \right) \right. \\ \left. - \cdots - nG \left(\frac{z_k}{\sqrt{n}} + \frac{t_{k-1}}{\sqrt{n}} \right) - nG \left(\frac{z_k}{\sqrt{n}} \right) \right\} \prod_{j=1}^{k-1} \mu_n(z_k + dt_j). \quad (11)$$

Now recall that the entropy function $H(\cdot)$ is real analytic in (θ, ζ) and therefore has a power series expansion $H(x) = \sum_{\nu=2}^{\infty} c_{\nu} x^{\nu}$, where $c_2 = \frac{1}{2}$ of course. By assumption (IV), $G(\cdot)$ must be of the form

$$G(x) = \sum_{\nu=2}^{[p]} c_{\nu} x^{\nu} - \lambda |x|^p + o(|x|^p) \quad \text{as } x \rightarrow 0. \quad (12)$$

We will use (12) to expand the function appearing inside the curly brackets in (11). A little elementary algebra shows that the sum of terms of this function arising from the term $\frac{1}{2}x^2$ in (12) is

$$-\frac{1}{2} \left[\left(1 - \frac{1}{k} \right) \sum_{j=1}^{k-1} t_j^2 - \frac{2}{k} \sum_{1 \leq i < j \leq k-1} t_i t_j \right] = -\frac{1}{2} \mathbf{t} \mathbf{M} \mathbf{t}', \quad (13)$$

where \mathbf{M} is the $(k-1) \times (k-1)$ matrix

$$\mathbf{M} = \begin{bmatrix} 1-1/k & -1/k & \cdots & -1/k \\ -1/k & 1-1/k & \cdots & -1/k \\ \vdots & & \ddots & \\ -1/k & -1/k & \cdots & 1-1/k \end{bmatrix}$$

and \mathbf{t} is the row vector $(t_1, t_2, \dots, t_{k-1})$. The sum of terms arising from the term $c_{\nu} x^{\nu}$ ($2 < \nu \leq [p]$) in (12) is

$$knc_{\nu} \left(\frac{z_k}{\sqrt{n}} + \frac{t_1 + \cdots + t_{k-1}}{k\sqrt{n}} \right)^{\nu} - nc_{\nu} \left(\frac{z_k}{\sqrt{n}} + \frac{t_1}{\sqrt{n}} \right)^{\nu} - \cdots - nc_{\nu} \left(\frac{z_k}{\sqrt{n}} + \frac{t_{k-1}}{\sqrt{n}} \right)^{\nu} - nc_{\nu} \left(\frac{z_k}{\sqrt{n}} \right)^{\nu} \\ = nc_{\nu} \left(\frac{z_k}{\sqrt{n}} \right)^{\nu} \left[k \left(1 + \frac{t_1 + \cdots + t_{k-1}}{kz_k} \right)^{\nu} - \left(1 + \frac{t_1}{z_k} \right)^{\nu} - \cdots - \left(1 + \frac{t_{k-1}}{z_k} \right)^{\nu} - 1 \right]. \quad (14)$$

The function in square brackets in (14) is $O(1/z_k^2)$ as $z_k \rightarrow \infty$, uniformly in $t_1 \in [u_1, v_1], \dots, t_{k-1} \in [u_{k-1}, v_{k-1}]$. (To see this replace $1/z_k$ by x : the resulting function $f(x)$ of x is $f(0) + f'(0)x + O(x^2) = O(x^2)$ as $x \rightarrow 0$. As a matter of fact $\frac{1}{2}f''(0) = \nu(\nu-1)[-\frac{1}{2}\mathbf{t}\mathbf{M}\mathbf{t}']$). Hence (14) is

$$nc_{\nu} (z_k/\sqrt{n})^{\nu} O(1/z_k^2) = O(z_k^{\nu-2} n^{1-\nu/2}).$$

The same applies if $\nu = p$, i.e. to terms arising from $-\lambda|x|^p$. Finally the sum of terms that are $o(|\cdot|^p)$ is $no((z_k/\sqrt{n})^p) = o(z_k^p n^{1-p/2})$. Putting these facts together and changing the variable in the outer integral of (10) by setting

$$z_k = yn^{1/2-1/p}$$

we see that (10) is equal to

$$\begin{aligned} \frac{\mathcal{Z}_n}{\mathcal{Z}_{kn}} \int_a^b \left[\int_{u_1}^{v_1} \cdots \int_{u_{k-1}}^{v_{k-1}} \exp\{-\tfrac{1}{2}t\mathbf{M}t' + B_n(y, t)\} \right. \\ \left. \cdot \prod_{j=1}^{k-1} \mu_n(yn^{1/2-1/p} + dt_j) \right] Q_n(dy) \end{aligned} \quad (15)$$

in the notation of (8), (9), where $B_n(y, t)$ is a sum of terms over ν , $2 < \nu \leq p$, which are of the form

$$O(y^{\nu-2} n^{(1/2-1/p)(\nu-2)+1-\nu/2}) = O(y^{\nu-2} n^{(2-\nu)/p})$$

if $2 < \nu < p$, or

$$o(y^p n^{(1/2-1/p)+1-p/2}) = o(y^p)$$

if $\nu = p$. It follows that $B_n(y, t)$ converges to 0 as $n \rightarrow \infty$, uniformly in $y \in [a, b]$, $t_1 \in [u_1, v_1], \dots, t_{k-1} \in [u_{k-1}, v_{k-1}]$.

The ratio

$$\mathcal{Z}_n / \mathcal{Z}_{kn} = (\mathcal{Z}_n / n^{1/2-1/p}) / (\mathcal{Z}_{kn} / (kn)^{1/2-1/p}) k^{1/2-1/p}$$

converges to $k^{-1/2+1/p}$. Hence, by Corollaries 1 and 2 and the Proposition in Section 2 of [12], (15) converges to

$$\begin{aligned} \frac{1}{k^{1/2-1/p}} \int_a^b \left[\int_{u_1}^{v_1} \cdots \int_{u_{k-1}}^{v_{k-1}} \exp\{-\tfrac{1}{2}t\mathbf{M}t'\} \right. \\ \left. \cdot \prod_{j=1}^{k-1} \frac{1}{\sqrt{2\pi}} \exp(-\lambda y^p) dt_j \right] \frac{\exp(-\lambda y^p) dy}{\int_{-\infty}^{\infty} \exp(-\lambda z^p) dz} \\ = k^{1/p} \int_a^b \left[(2\pi)^{-(k-1)/2} k^{-1/2} \int_{u_1}^{v_1} \cdots \int_{u_{k-1}}^{v_{k-1}} \exp\{-\tfrac{1}{2}t\mathbf{M}t'\} \prod_{j=1}^{k-1} dt_j \right] \\ \cdot \frac{\exp(-k\lambda y^p) dy}{\int_{-\infty}^{\infty} \exp(-\lambda z^p) dz}. \end{aligned} \quad (16)$$

The $(k-1)$ -dimensional distribution with probability density $\text{const.} \exp(-\tfrac{1}{2}t\mathbf{M}t')$ is the $(k-1)$ -dimensional normal distribution with mean vector 0 and $(k-1) \times (k-1)$ covariance matrix

$$\Gamma = \begin{bmatrix} 2 & 1 & \cdots & 1 \\ 1 & 2 & & 1 \\ \vdots & & \ddots & \\ 1 & 1 & \cdots & 2 \end{bmatrix}$$

with $\det \Gamma = k$. We have therefore proved the following theorem.

Theorem 3. *The asymptotic distribution of $S_n^k / n^{1-1/p}$ has probability density function $\text{const.} \exp(-k\lambda |x|^p)$, $-\infty < x < \infty$. The random vector*

$$\left(\frac{S_n^1}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}}, \frac{S_n^2}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}}, \dots, \frac{S_n^{k-1}}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}} \right)$$

is asymptotically independent of $S_n^k/n^{1-1/p}$, and its asymptotic distribution is the same as that of

$$(Y_1 - Y_k, Y_2 - Y_k, \dots, Y_{k-1} - Y_k),$$

where Y_1, Y_2, \dots, Y_k are independent standard normal random variables. \square

An immediate consequence is that $S_n^1/n^{1-1/p}, S_n^2/n^{1-1/p}, \dots, S_n^k/n^{1-1/p}$ are asymptotically equal, in the sense that the difference of any two of them converges to 0 in probability. (See [12] for background.) Since $S_n^1 + S_n^2 + \dots + S_n^k = S_{kn}$, we deduce that $S_n^k/n^{1-1/p}$ and $k^{-1}S_{kn}/n^{1-1/p} = k^{-1/p}S_{kn}/(kn)^{1-1/p}$ are asymptotically equal.

Corollary 3. *The random vector*

$$\left(\frac{S_n^1}{\sqrt{n}} - \frac{1}{k} \frac{S_{kn}}{\sqrt{n}}, \frac{S_n^2}{\sqrt{n}} - \frac{1}{k} \frac{S_{kn}}{\sqrt{n}}, \dots, \frac{S_n^k}{\sqrt{n}} - \frac{1}{k} \frac{S_{kn}}{\sqrt{n}} \right)$$

is asymptotically independent of $S_{kn}/(kn)^{1-1/p}$, and its asymptotic distribution is that of

$$\left(Y_1 - \frac{1}{k} \sum_{i=1}^k Y_i, Y_2 - \frac{1}{k} \sum_{i=1}^k Y_i, \dots, Y_k - \frac{1}{k} \sum_{i=1}^k Y_i \right)$$

where Y_1, \dots, Y_k are as in Theorem 3. \square

This follows from Theorem 3 and the fact that for $j = 1, 2, \dots, k$,

$$\frac{S_n^j}{\sqrt{n}} - \frac{1}{k} \frac{S_{kn}}{\sqrt{n}} = \left(\frac{S_n^j}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}} \right) - \frac{1}{k} \sum_{i=1}^k \left(\frac{S_n^i}{\sqrt{n}} - \frac{S_n^k}{\sqrt{n}} \right).$$

This corollary establishes for general $G(\cdot)$, subject to assumptions (I) to (IV), the convergence of finite-dimensional distributions to those of a Brownian bridge asserted by Theorem 1 for the special case $G(x) = \frac{1}{2}x^2$, since

$$\frac{S_n^j}{n^{1-1/p}} = \frac{S_{kn}}{n^{1-1/p}} \frac{1}{k} + \frac{1}{n^{1/2-1/p}} \left(\frac{S_n^j}{\sqrt{n}} - \frac{1}{k} \frac{S_{kn}}{\sqrt{n}} \right).$$

The only thing left for a full analogue of Theorem 1 is to establish the tightness of the sequence of corresponding distributions in the space of continuous paths, but we will not go into this routine matter here. We will instead give a more geometric formulation of Theorem 3 in terms of the distribution of the random vector $V_n = (S_n^1, S_n^2, \dots, S_n^k)$ in \mathbb{R}^k . Set $d = (1, 1, \dots, 1) \in \mathbb{R}^k$ and let $\{e_2, e_3, \dots, e_k\}$ be an orthonormal base in the $(k-1)$ -dimensional linear subspace of \mathbb{R}^k orthogonal to d .

Theorem 4. *Let*

$$V_n = \frac{1}{k} S_{kn} d + U_n^2 e_2 + U_n^3 e_3 + \dots + U_n^k e_k$$

be the representation of V_n in terms of the base $\{\mathbf{d}, \mathbf{e}_2, \mathbf{e}_3, \dots, \mathbf{e}_k\}$ in \mathbb{R}^k . The asymptotic joint distribution of

$$\frac{1}{k} \frac{S_{kn}}{n^{1-1/p}}, \frac{U_n^2}{\sqrt{n}}, \frac{U_n^3}{\sqrt{n}}, \dots, \frac{U_n^k}{\sqrt{n}},$$

as $n \rightarrow \infty$, is that of k independent random variables Z_1, Z_2, \dots, Z_k , where Z_1 has probability density function $\text{const. exp}(-k\lambda|x|^p)$, $-\infty < x < \infty$, and Z_2, Z_3, \dots, Z_k are standard normal. \square

In the context of [3] the direction of \mathbf{d} is the direction of degeneracy: if V_n is scaled by the factor $n^{-(1-1/p)}$, then its asymptotic distribution is carried by the one-dimensional subspace generated by \mathbf{d} . To get a genuinely k -dimensional asymptotic distribution one must scale by the factor $n^{-(1-1/p)}$ in the direction of \mathbf{d} and by the factor $n^{-1/2}$ in all directions orthogonal to \mathbf{d} . Theorem 4 is an immediate consequence of Theorem 3. If for instance

$$\begin{aligned} \mathbf{e}_2 &= \left(\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}, 0, 0, \dots, 0 \right), \\ \mathbf{e}_3 &= \left(\frac{1}{\sqrt{6}}, \frac{1}{\sqrt{6}}, -\frac{2}{\sqrt{6}}, 0, \dots, 0 \right), \\ &\vdots \\ \mathbf{e}_k &= ((k^2 - k)^{-1/2}, \dots, (k^2 - k)^{-1/2}, -(k-1)(k^2 - k)^{-1/2}), \end{aligned}$$

then

$$\frac{U_n^2}{\sqrt{n}}, \frac{U_n^3}{\sqrt{n}}, \dots, \frac{U_n^k}{\sqrt{n}}$$

converge in distribution to

$$\frac{1}{\sqrt{2}}(Y_1 - Y_2), \frac{1}{\sqrt{6}}(Y_1 + Y_2 - 2Y_3), \dots, (k^2 - k)^{-1/2}(Y_1 + \dots + Y_{k-1} - (k-1)Y_k)$$

in the notation of Theorem 3.

References

- [1] R. Azencott, Grandes déviations et applications, in: Ecole d'Été de Probabilités de Saint-Flour VIII, Lecture Notes in Math., Vol. 774 (Springer, Berlin, 1980) pp. 1–176.
- [2] G. Ben Arous and M. Brunaud, Méthode de Laplace: Etude variationnelle des fluctuations de diffusions de type "champ moyen", Stochastics and Stochastics Reports 31 (1990) 79–144.
- [3] E. Bolthausen, Laplace approximations for sums of independent random vectors, Part II: Degenerate maxima and manifolds of maxima, Probab. Theory Rel. Fields 76 (1987) 167–206.
- [4] N.R. Chaganty and J. Sethuraman, Limit theorems in the area of large deviations for some dependent random variables, Ann. Probab. 15 (1987) 628–645.

- [5] D.A. Dawson, Critical dynamics and fluctuations for a mean-field model of cooperative behaviour, *J. Statist. Phys.* 31 (1983) 29–85.
- [6] F. Dunlop and C.M. Newman, Multicomponent field theories and classical rotators, *Commun. Math. Phys.* 44 (1975) 223–235.
- [7] R.S. Ellis and C.M. Newman, Limit theorems for sums of dependent random variables occurring in statistical mechanics, *Z. Wahrsch. Verw. Gebiete* 44 (1978) 117–139.
- [8] R.S. Ellis, *Entropy, Large Deviations and Statistical Mechanics* (Springer, New York, 1985).
- [9] W. Feller, *An Introduction to Probability Theory and its Applications*, Vol. II (Wiley, New York, 1971, 2nd ed.).
- [10] M. Kac, Mathematical mechanisms of phase transition, in: M. Chretien, E.P. Gross and S. Deser, eds., *Statistical Physics: Phase Transitions and Superfluidity*, Vol. 1, Brandeis University Summer Institute in Theoretical Physics (Gordon and Breach, New York, 1968) pp. 241–305.
- [11] A. Martin-Löf, A Laplace approximation for sums of independent random variables, *Z. Wahrsch. Verw. Gebiete* 59 (1982) 101–115.
- [12] F. Papangelou, On the Gaussian fluctuations of the critical Curie–Weiss model in statistical mechanics, *Probab. Theory Rel. Fields* 83 (1989) 265–278.
- [13] V.V. Petrov, Sums of Independent Random Variables, *Ergeb. Math. Grenzgeb.*, Vol. 82 (Springer, Berlin, 1975).
- [14] B. Simon and R.B. Griffiths, The $(\varphi^4)_2$ field theory as a classical Ising model, *Commun. Math. Phys.* 33 (1973) 145–164.
- [15] C.J. Thompson, *Classical Equilibrium Statistical Mechanics* (Clarendon, Oxford, 1988).